

today: • themes
• prob. review

key themes:

I) representation: how to represent structured prob. dist.?

(probability)

• graph $\leadsto$ factorization
• parameterization $\longrightarrow$ full table
vs.
"exponential family"

II) estimation (statistics): given data, how to learn/estimate the parameters of dist.

$\leadsto$ learning e.g. maximum likelihood
max. entropy
moment matching

III) "probabilistic inference" (CS)

: answer questions about data

e.g. compute $p(y|x)$ or $p(x)$
query observation

$\leadsto$ computation e.g. • message passing
• approximate inference
- sampling
- variational methods

probability review

why? $\leadsto$ principled framework to model uncertainty

sources of uncertainty

1) intrinsic uncertainty $\leadsto$ quantum mechanics

2) partial information / observation: • card game
• rolling a die
 $\leadsto$ don't know the initial conditions "exactly enough"

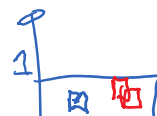
3) incomplete modeling of a complex phenomenon

(computational issues are also important)

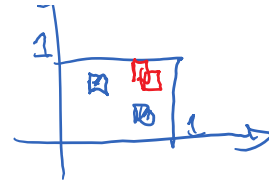
example: "most bird can fly"
 $\leadsto$ simple rule can be advantageous but then yields uncertainty

• "AI"

probabilities $\leadsto$ like areas



probabilities $\rightarrow$ like areas



Is that "aleatoric"?

Notation: $X_1, X_2, X_3, \dots, X, Y, Z$
 $\leftarrow$ random variables (usually real-valued)

$x_1, x_2, x_3, \dots, x, y, z$
 $\leftarrow$ their realizations

X is a random variable $\rightarrow$ represents an uncertain quantity

eg. X
 "result of a roll of a die"

$X=x$ represents the "event" that X takes the value x

$\Omega \rightarrow$ the sample space of "elementary events"
 possible values of my R.V.

$$\Omega = \{1, 2, 3, 4, 5, 6\}$$

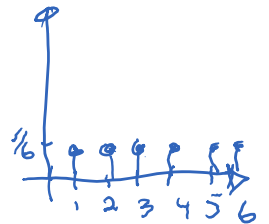
two types of R.V. $\leftarrow$ discrete where Ω is countable
continuous " Ω " uncountable

I) [assume Ω is countable $\rightarrow$ discrete case]

discrete R.V. is characterized by a probability mass function (pmf)

"lower case" $\rightarrow p(x)$ for $x \in \Omega$

$$\text{pmf } p \text{ is s.t. } \begin{cases} p(x) \geq 0 \forall x \in \Omega \\ \sum_{x \in \Omega} p(x) = 1 \end{cases}$$



probability distribution "big" P

is a mapping $P: \mathcal{E} \rightarrow [0, 1]$

that satisfies the Kolmogorov axioms

$$P(\Omega) = 1$$

$$P\left(\bigcup_{i=1}^{\infty} E_i\right) = \sum_{i=1}^{\infty} P(E_i) \text{ when } E_i \text{'s are mutually disjoint}$$

$\mathcal{E} = 2^{\Omega} \triangleq$ set of all subsets of Ω

set of "events"

(" σ -field" in measure theory)

crucial when Ω is uncountable

relation: $P(\{X=x\}) = p(x)$

"formal prob world"

$$\{X=x\}$$

$$\{\omega: X(\omega)=x\}$$

$$P(\{X=x \text{ or } X=x'\}) = P(\{X=x\}) + P(\{X=x'\}) = p(x) + p(x')$$

$$E = \{x, x'\}$$

(if $x \neq x'$)

$$\{X=x\} \triangleq \{\omega: X(\omega)=x\}$$

$E = \{x, x'\}$
for discrete R.V.

$$P(E) = \sum_{x \in E} p(x)$$

(if $x \neq x'$)

(5h30)

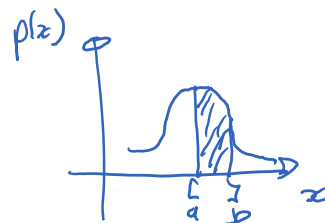
continuous R.V.

continuous R.V. is characterized by a probability density function p (pdf)

$$p: \Omega \rightarrow \mathbb{R}$$

$$p(x) \geq 0 \quad \forall x \in \Omega$$

$$p \text{ is integrable and } \int_{\Omega} p(x) dx = 1$$



prb. of events

$$\Omega = \mathbb{R} : P([a, b]) = \int_a^b p(x) dx$$

$\mathcal{E} \rightarrow$ Borel σ -fields

$$p(x) = \lim_{n \rightarrow \infty} \frac{P(B_n)}{\text{vol}(B_n)}$$

where B_n are balls centered around x with decreasing radius

reap: discrete R.V. X ; pmf $p(x)$ $\Leftrightarrow P\{X=x\} = p(x)$

cts. R.V. X ; pdf $p(x)$ $P\{X=x\} = 0$

$$P\{X \in x \pm \frac{dx}{2}\} \approx p(x) dx$$

$$P\{X \in x \pm \frac{\Delta}{2}\} = \int_{x-\frac{\Delta}{2}}^{x+\frac{\Delta}{2}} p(u) du$$

[side note: can change pdf p on a countable # of points]

without changing anything

has $\uparrow$ measure zero according to Lebesgue measure

joint, marginal, etc. (context: multivariate R.V.)
"random vector"

$$Z = (X, Y)$$

(discrete) pmf of Z

$$\Omega_Z \triangleq \Omega_X \times \Omega_Y$$

"joint pmf" on $X \times Y$

$$p(x, y) = P\{X=x, Y=y\}$$

Joint distribution

can represent elementary events of $\Omega(x,y)$ as table

e.g.:

	$X=0$	$X=1$
$Y=0$	0	$1/2$
$Y=1$	$1/2$	0

$p(1,1)$

say X is whether die result is even
 " Y " " " " is odd

cts. case: if X & Y are cts. $P\{X \in \text{box}, Y \in \text{region}\} = \int_{\text{box}} \int_{\text{region}} p(x,y) dx dy$ (joint pdf)

marginal distribution (in context of a joint dist.)

↳ dist. of a component of a random vector

$$P\{X=x\} = \sum_{y \in \Omega_Y} \underbrace{P\{X=x, Y=y\}}_{p(x,y)} \quad \text{"sum rule"}$$

"marginalizing out Y "
 cts. $\rightarrow$ integrate

cts. case joint pdf $p(x,y)$

$$p(x) = \int_y p(x,y) dy$$

mixed type: X cts.
 Y discrete

$p(x,y)$
 discrete $\rightarrow$ sum

$$P\{X \in \text{set}\} = \sum_y \int_{x \in \text{set}(y)} p(x,y) dx$$

independence

$X \text{ is independent of } Y \Leftrightarrow p(x,y) = p(x)p(y) \quad \forall (x,y) \in \Omega_X \times \Omega_Y$

notation: $X \perp Y$

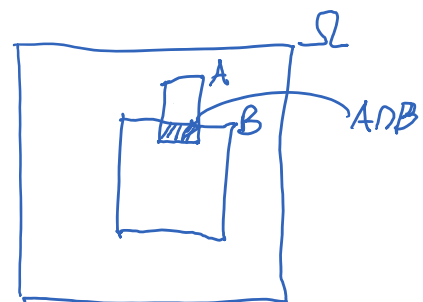
R.V. $X_1, \dots, X_n$ are "mutually independent" $\Leftrightarrow p(x_{1:n}) = \prod_{i=1}^n p(x_i)$
 $\forall x_{1:n} \in \bigtimes_{i=1}^n \Omega_{X_i}$

conditioning:

for events A & B , suppose $P(B) \neq 0$

then $P(A|B) \triangleq \frac{P(A \cap B)}{P(B)}$

Normalizer



$$P(B) = \sum P(A \cap B)$$

for discrete R.V.

define "conditional pmf"

$$p(x|y) \triangleq P(\{X=x\} | \{Y=y\}) \triangleq \frac{P(\{X=x, Y=y\})}{P(\{Y=y\})} = \frac{p(x,y)}{p(y)} \quad \begin{array}{l} \text{joint pmf} \\ p(x,y) \\ \text{marginal pmf} \\ p(y) \end{array}$$

 $p(x|y)$ or $p(x,y)$
 $\uparrow$
 "proportional to"
normalization
const. $\sum_x p(x,y) = p(y)$

for cts. R.V.

"conditional pdf" $p(x|y) \triangleq \frac{p(x,y)}{p(y)}$

subtle point: $p(x|y)$ is undefined when $p(y)=0$ see Borel-Kolmogorov paradox for weird properties of cond. prob.[Borel-Kolmogorov paradox on wikipedia](#)

* note: independent $\times \parallel \vee$; $p(x|y) = \frac{p(x,y)}{p(y)} \stackrel{\text{by II}}{=} \frac{p(x)p(y)}{p(y)} = p(x)$